

# Unconditional Secrecy and Computational Complexity against Wireless Eavesdropping

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# Introduction

- Secrecy unconditional on Eve's channel condition (such as Eve's number of antennas and Eve's noise level) is crucial for many applications.
- Unconditional secrecy (UNS) can be achieved by secret key generation (SKG) utilizing user's reciprocal channel state information (CSI) being independent of Eve's CSI.
- But for (direct) secret information transmission (SIT) between users, there has been little research activity to address UNS.
- In this paper, we evaluate the UNS of several prior methods for SIT and also examine the computational complexities that they impose onto Eve if Eve (with a superior channel condition) wants to break any further secrecy beyond UNS.
- We also show a study of UNS and eavesdropping complexity of a recently proposed method called randomized reciprocal channel modulation (RRCM).

# Conventional MIMO Beamforming

- Consider a MIMO channel from Alice to Bob

$$\mathbf{y}_B(k) = \mathbf{H}\mathbf{x}_A(k) + \mathbf{w}_B(k)$$

where  $\mathbf{H} \in \mathbb{C}^{N_B \times N_A}$  is the reciprocal channel matrix known to both Alice and Bob (via two-way training) but unknown to Eve.

- Both Alice and Bob know the SVD  $\mathbf{H} = \sum_{i=1}^r \sigma_i \mathbf{u}_i \mathbf{v}_i^H = \mathbf{U}\mathbf{\Sigma}\mathbf{V}^H$  with  $r = \min(N_A, N_B)$ , and Alice transmits

$$\mathbf{x}_A(k) = \mathbf{V}\mathbf{c}_A(k)$$

where  $\mathbf{c}_A(k) \in \mathbb{C}^{r \times 1}$  for all  $k$  are symbol vectors. We know that Bob can successfully decode all information in  $\mathbf{c}_A(k)$  for all  $k$  provided that the data rate in the  $i$ th element of  $\mathbf{c}_A(k)$  is less than the capacity of the  $i$ th subchannel.

## Conventional MIMO Beamforming (Cont.)

- The signal received by Eve with negligible noise is

$$\mathbf{y}_E(k) = \mathbf{G}_A \mathbf{x}_A(k)$$

where  $\mathbf{G}_A \in \mathbb{C}^{N_E \times N_A}$  is known to Eve (due to training pilot from Alice) but unknown to Alice and Bob. Eve with  $N_E \geq N_A$  is able to recover  $\mathbf{x}_A(k) = \mathbf{V} \mathbf{c}_A(k)$  for all  $k$ . But without knowing  $\mathbf{V}$ , Eve is unable to retrieve all information from  $\mathbf{c}_A(k)$ .

- However, among all possible random guesses of  $\mathbf{c}_A(k)$  for  $1 \leq k \leq r$  (for example), there is a correct one. With this correct guess, Eve knows a correct choice of  $\mathbf{V}$  by solving the linear equations  $\mathbf{x}_A(k) = \mathbf{V} \mathbf{c}_A(k)$  for  $1 \leq k \leq r$ . Therefore,  $\mathbf{x}_A(k)$  for any  $k > r$  no longer contains further UNS from Eve.
- In other words, the strict UNS of the above scheme in each coherence period is no more than the entropy of  $r^2$  symbols in  $\mathbf{c}_A(k)$  for  $1 \leq k \leq r$ , and the complexity for Eve to obtain  $\mathbf{V}$  mainly involves a set of linear equations, which is in the order of  $\mathcal{O}(N_A r^2)$ .

# Randomized MISO Beamforming

- Li et al introduced a randomized MISO beamforming for a MISO user channel where  $N_A > N_B = 1$ . The user's channel is described by

$$y_B(k) = \mathbf{h}^T \mathbf{x}_A(k) + \mathbf{w}_B(k)$$

where  $\mathbf{h}$  is known to Alice (through a pilot from Bob) but unknown to Bob,  $\mathbf{x}_A(k) = \mathbf{w}_k c_A(k)$  and  $\mathbf{w}_k \in \mathbf{C}^{N_A \times 1}$  is randomly chosen for each  $k$  subject to

$$\mathbf{h}^T \mathbf{w}_k = \|\mathbf{h}\|.$$

- With  $c_A(1)$  (for example) as a training symbol (i.e., known to all), Bob can obtain  $\|\mathbf{h}\|$  and hence decode the information in  $c_A(k)$  for all  $k > 1$  from

$$y_B(k) = \|\mathbf{h}\| c_A(k) + \mathbf{w}_B(k).$$

## Randomized MISO Beamforming (Cont.)

- At the same time, Eve with  $N_E \geq N_A$  antennas and negligible noise receives

$$\mathbf{y}_E(k) = \mathbf{G}_A \mathbf{x}_A(k) = \mathbf{G}_A \mathbf{w}_k c_A(k)$$

where  $\mathbf{G}_A \in \mathbb{C}^{N_E \times N_A}$  is unknown to Eve (or anyone else). With unknown  $\mathbf{G}_A \mathbf{w}_k$ , Eve is unable to decode all information in  $c_A(k)$ .

- But if Eve has guessed  $c_A(2), \dots, c_A(N_A)$  correctly (in addition to the known  $c_A(1)$ ), then Eve can compute a vector  $\mathbf{q} \in \mathbb{C}^{N_E \times 1}$  such that  $\mathbf{q}^H \mathbf{y}_E(k) = c_A(k)$  for  $1 \leq k \leq N_A$  or equivalently  $\mathbf{q}^H \mathbf{G}_A \mathbf{w}_k = 1$  for  $1 \leq k \leq N_A$ . Assuming that  $\mathbf{w}_1, \dots, \mathbf{w}_{N_A}$  are linearly independent of each other,  $\mathbf{q}^H \mathbf{G}_A$  is unique and hence equals to  $\mathbf{h}^T \frac{1}{\|\mathbf{h}\|}$ . Then Eve can use the same  $\mathbf{q}$  to obtain  $\mathbf{q}^H \mathbf{y}_E(k) = c_A(k)$  for all  $k > N_A$ .
- Therefore, the strict UNS of the above scheme is no more than the entropy of  $N_A - 1$  symbols from Alice, and the complexity for Eve to obtain the equalization vector  $\mathbf{q}$  is (easy to prove) in the order of  $\mathcal{O}(N_E N_A^2)$ .

## Artificial Noise from Multi-Antenna Transmitter

- Using the artificial noise idea from Negi and Goel, a multi-antenna Alice can transmit

$$\mathbf{x}_A(k) = \mathbf{V}_1 \mathbf{s}(k) + \mathbf{V}_2 \mathbf{n}(k)$$

where  $\mathbf{s}(k) \in \mathbb{C}^{r_1 \times 1}$  with  $r_1 \leq r$  is a signal vector,  $\mathbf{n}(k) \in \mathbb{C}^{(N_A - r_1) \times 1}$  is an artificial noise meant to jam Eve,  $\mathbf{V}_1 = [\mathbf{v}_1, \dots, \mathbf{v}_{r_1}]$  and  $\mathbf{V}_2$  consists of  $N_A - r_1$  orthogonal complement vectors of  $\mathbf{V}_1$ .

- The signal received by Bob is

$$\begin{aligned} \mathbf{y}_B(k) &= \mathbf{H} \mathbf{x}_A(k) + \mathbf{w}_B(k) \\ &= \mathbf{U}_1 \mathbf{\Sigma}_1 \mathbf{s}(k) + \mathbf{U}_2 \mathbf{\Sigma}_2 \mathbf{n}(k) + \mathbf{w}_B(k) \end{aligned}$$

where  $\mathbf{U}_1$ ,  $\mathbf{U}_2$ ,  $\mathbf{\Sigma}_1$  and  $\mathbf{\Sigma}_2$  are corresponding partitions of  $\mathbf{U}$  and  $\mathbf{\Sigma}$ . Because of  $\mathbf{U}_1^H \mathbf{U}_2 = 0$ , the artificial noise  $\mathbf{n}(k)$  has zero impact on Bob's ability to decode the information from Alice.

- Note that for Bob to estimate  $\mathbf{s}(k)$  from  $\mathbf{y}_B(k)$ , Bob first needs to know  $\mathbf{U}_1 \mathbf{\Sigma}_1$ , which requires (for example)  $\mathbf{s}(k)$  for  $k = 1, \dots, r_1$  to be pilot vectors from Alice.

## Artificial Noise from Multi-Antenna Transmitter (Cont.)

- Now consider Eve with  $N_E$  antennas and negligible (self) noise. Corresponding to  $\mathbf{x}_A(k)$  from Alice, the signal received by Eve is

$$\mathbf{y}_E(k) = \mathbf{G}_A \mathbf{V}_1 \mathbf{s}(k) + \mathbf{G}_A \mathbf{V}_2 \mathbf{n}(k)$$

where  $\mathbf{G}_A$ ,  $\mathbf{V}_1$  and  $\mathbf{V}_2$  are all unknowns to Eve.

- But in multipath-rich environment, the entries in  $\mathbf{G}_A$  can be modelled to be i.i.d. with zero mean and variance  $\sigma_G^2$ . Then with  $N_E \gg 1$  and  $N_E \geq N_A$ , we have  $\frac{1}{N_E} \mathbf{G}_A^H \mathbf{G}_A \approx \sigma_G^2 \mathbf{I}_{N_A}$ , and hence  $\mathbf{G}_A \mathbf{V}_1$  and  $\mathbf{G}_A \mathbf{V}_2$  have approximately orthogonal ranges.

## Artificial Noise from Multi-Antenna Transmitter (Cont.2)

- Since  $\mathbf{s}(1), \dots, \mathbf{s}(r_1)$  are known pilots, Eve can compute  $\mathbf{Q} \in \mathbb{C}^{N_E \times r_1}$  such that

$$\mathbf{Q}^H \mathbf{y}_E(k) \approx \mathbf{s}(k) \quad (1)$$

for  $1 \leq k \leq r_1$ , or  $\sum_{k=1}^{r_1} \|\mathbf{Q}^H \mathbf{y}_E(k) - \mathbf{s}(k)\|^2$  is minimized.

- For large  $N_E$ , there exists a  $\mathbf{Q} \in \text{range}(\mathbf{G}_A \mathbf{V}_1)$  such that  $\mathbf{Q}^H \mathbf{G}_A \mathbf{V}_1 \approx \mathbf{I}_{r_1}$  and  $\mathbf{Q}^H \mathbf{G}_A \mathbf{V}_2 \approx 0$  and hence (1) holds.
- The solution space of  $\mathbf{Q}$  to (1) may be large due to large  $N_E$ . But Eve can choose the minimum-norm solution given by  $\mathbf{Q}^H = \mathbf{S}(\mathbf{Y}_E^H \mathbf{Y}_E)^{-1} \mathbf{Y}_E$  where  $\mathbf{S} = [\mathbf{s}(1), \dots, \mathbf{s}(r_1)]$  and  $\mathbf{Y}_E = [\mathbf{y}_E(1), \dots, \mathbf{y}_E(r_1)]$ .
- Therefore, the strict UNS of the artificial noise scheme is zero, and the complexity for Eve to obtain an accurate equalizer  $\mathbf{Q}$  is in order of  $\mathcal{O}(r_1^2 N_E)$ .

# Randomized Reciprocal Channel Modulation (RRCM)

- Consider a MISO user channel with  $N_A = n_A^2 \geq 4$  and  $N_B = 1$ . Using a pilot from Bob, Alice obtains an estimate of  $\mathbf{h} = [h_1, \dots, h_{N_A}]^T$ .
- Then, Alice computes  $\mathbf{D}_s = \text{diag}[m_{s,1}, \dots, m_{s,N_A}]$  for  $1 \leq s \leq S$  as follows. Define  $\mathbf{H}_s \in \mathbb{C}^{n_A \times n_A}$  with  $(\mathbf{H}_s)_{i,l} = h_{(i-1)n_A+l} m_{s,(i-1)n_A+l}$ . Denote the SVD of  $\mathbf{H}_s$  as

$$\mathbf{H}_s = \sum_{i=1}^{n_A} \sigma_{i,s} \mathbf{u}_{i,s} \mathbf{v}_{i,s}^H = \mathbf{U}_s \mathbf{\Sigma}_s \mathbf{V}_s^H \quad (2)$$

where the first element of the vector  $\mathbf{u}_{i,s}$  is normalized to be real.

- Also let

$$r_s = \sigma_{1,s} e^{j\mu_{1,s}} \quad (3)$$

where  $\mu_{1,s}$  is the phase of the first element of  $\mathbf{v}_{1,s}$ . For each  $s$ , Alice chooses a sufficiently random  $r_s$  to hide the information of  $c_s$  in  $r_s c_s$ , and chooses randomly all other components in  $\mathbf{U}_s$ ,  $\mathbf{\Sigma}_s$  and  $\mathbf{V}_s$ . Then Alice determines  $\mathbf{D}_s = \text{diag}[m_{s,1}, \dots, m_{s,N_A}]$  from  $\mathbf{H}_s$ .

## RRCM (Cont.)

- Then, Alice sends a pure and several randomized pilots  $\sqrt{P_T}\mathbf{I}_{N_A}, \sqrt{P_T}\mathbf{D}_1\mathbf{I}_{N_A}, \dots, \sqrt{P_T}\mathbf{D}_S\mathbf{I}_{N_A}$  so that Bob can obtain  $\mathbf{h}$  and all entries in  $\mathbf{H}_s$ . Hence, Bob can use (2) and (3) to compute  $r_s$  for  $1 \leq s \leq S$ .
- Following the randomized pilots, Alice sends  $\sqrt{P_T}r_s c_s$  for  $1 \leq s \leq S$  from the antenna corresponding to the strongest channel, and then Bob receives

$$y_{B,s} = \sqrt{P_T}h_{max}r_s c_s + w_{B,s}$$

where  $h_{max} = \arg \max_{h_i} |h_i|$ . All channel estimation errors (if not too large) can be lumped into  $w_{B,s}$ . Since Bob knows  $\mathbf{h}$  and  $r_s$ , Bob can decode all information in  $c_s$  for all  $s$  (assuming that the information rate in  $c_s$  is so controlled that the probability of detection error is negligible).

## RRCM (Cont.2)

- Now consider Eve with  $N_E \geq 2$  and negligible noise. Due to the pilots from Alice, Eve knows its receive channel matrix  $\mathbf{G}_A$  and also  $\mathbf{G}_A \mathbf{D}_s$  for all  $s$ . Hence, Eve knows  $m_{s,i}$  for all  $s$  and  $i$ .
- Corresponding to the information symbols from Alice, Eve receives  $\mathbf{y}_{E,s} = \mathbf{g}_A r_s c_s$  where  $\mathbf{g}_A$  is one of the  $N_A$  columns in  $\mathbf{G}_A$  and can be identified by Eve. Consequently, Eve knows  $r_s c_s$  for all  $s$ . In order to decode the information in  $c_s$ , Eve must first determine  $r_s$ .
- Assume that Eve has guessed correctly  $c_s$  for  $1 \leq s \leq S_0$  and hence knows  $r_s$  for  $1 \leq s \leq S_0$ . In order to determine  $r_s$  for  $s > S_0$ , Eve now must determine  $\mathbf{h}$  using  $r_s$  for  $1 \leq s \leq S_0$  via (2) along with the conditions  $\mathbf{U}_s^H \mathbf{U}_s = \mathbf{I}_{n_A}$  and  $\mathbf{V}_s^H \mathbf{V}_s = \mathbf{I}_{n_A}$ .
- One can verify that the total number of real unknowns is

$$N_{unk} = 2n_A^2 + 2(n_A^2 - 1)S_0$$

and the total number of effective real equations is

$$N_{equ} = 2n_A^2 S_0.$$

## RRCM (Cont.3)

- For a finite number of solutions of  $\mathbf{h}$ , it is necessary (but not sufficient) that  $N_{unk} \leq N_{equ}$  or equivalently  $S_0 \geq n_A^2$ .
- Hence, the strict amount of UNS of RRCM is no less than the entropy of  $N_A = n_A^2$  symbols from Alice.
- Note that (2) is nonlinear. If Eve uses exhaustive search to find  $\mathbf{h}$ , Eve has to compute the  $n_A \times n_A$  SVD for each choice of  $\mathbf{h}$ . With  $N_q$  to be the number of quantization levels for each real element in  $\mathbf{h}$ , the number of these choices is in the order of  $\mathcal{O}(N_q^{2n_A^2})$ .
- Alternatively, Eve may apply the Newton's method to search for  $\mathbf{h}$  as shown in the paper. The complexity per iteration of the Newton's algorithm is in the order of  $\mathcal{O}(N_{unk}^3)$  but there are local extrema due to nonlinearity.

## Using the Newton's Method

- For the case of  $N_A = 4$ , we found that the Newton's method using known  $r_s$  for  $s = 1, \dots, S_0$  and  $S_0 = 4$  yielded correct solutions of  $\mathbf{h}$  from 94% of 100 random initializations of  $\mathbf{h}$ . We also found that the Newton's method with  $S_0 = 5$  yielded a correct solution of  $\mathbf{h}$  from each of 100 random initializations.
- For the case of  $N_A = 4$ , we also considered a phase-only modulation where  $r_s = e^{j\mu_{1,s}}$ . In this case, using  $r_s$  for  $s = 1, \dots, S_0$  and  $S_0 \leq 7$ , the Newton's method did not find any correct solution of  $\mathbf{h}$ . With  $S_0 = 8$ , the Newton's method yielded a correct solution of  $\mathbf{h}$  for only 1 out of 500 random initializations. Furthermore, we found that the Newton's method has a very poor convergence property for the phase-only modulation.
- For the case of  $N_A = 9$ , the necessary condition for a finite number of solutions of  $\mathbf{h}$  is now  $S_0 \geq 9$ . Using  $S_0 = 9$ , the Newton's method with 1000 random initializations of  $\mathbf{h}$  did not converge to any reasonable solution of  $\mathbf{h}$ . Increasing  $S_0$  did not help either.

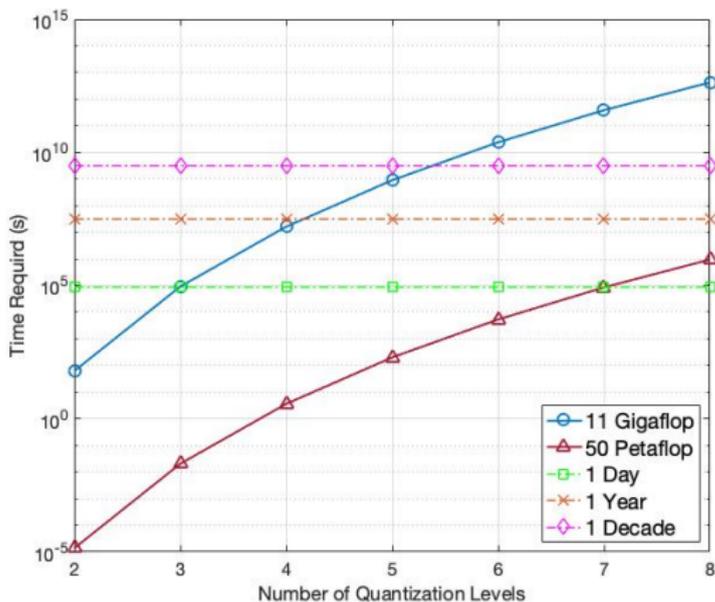
## Using Exhaustive Search

- We considered the case of  $N_A = 9$ . To obtain an estimate of how the required computational time varies with  $N_q$  (the number of quantization levels of each real element in  $\mathbf{h}$ ), we used our PC with 11.1 Gigaflops to compute the  $3 \times 3$  SVD (2) for all realizations of  $\mathbf{h}$  with  $N_q = 2$  and for  $s = 1, \dots, 9$ . We recorded this time as  $T_2$ .
- Then the required time for  $N_q$  is estimated by

$$T_{N_q} = \frac{N_q^{18}}{2^{18}} T_2$$

which is illustrated in the next Figure.

# Using Exhaustive Search (Cont.)



**Figure:** Computation time required for exhaustive search. Also shown in this figure is the time required if a supercomputer with 50 Petaflops is applied.

# Conclusion

Schemes:	MIMO-BF	R-MISO-BF	Artificial-N	RRCM
Symbols with UNS:	$r^2$ (see below)	$N_A - 1$	0	$N_A$
Complexity:	$N_A r^2$	$N_E N_A^2$	$N_E r_1^2$	NP

**Table:** Comparison of UNS and complexities where NP stands for “nondeterministic polynomial”,  $r \leq \min(N_A, N_B)$ ,  $r_1 \leq \min(r, N_A - 1)$ ,  $N_E \geq N_A$  for R-MISO-BF and Artificial-N, and  $N_E \gg 1$  for Artificial-N.

Note that if Alice uses pilot in beamspace for channel training, the UNS of MIMO-BF would be zero.

Thank You!